

# Active Learning 2 Final Presentation

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# 01

## Topic & Repository

# Topic




E Scooter - Helmet Problem

# Repository

## How? -> object detection

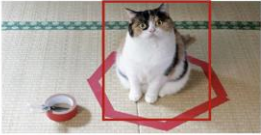
1



Is this image of Cat or not?

Image classification problem

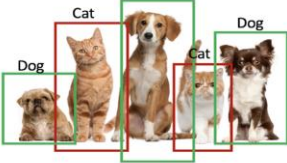
2



Where is Cat?

Classification with localization problem





3



Which animals are there in image and where?

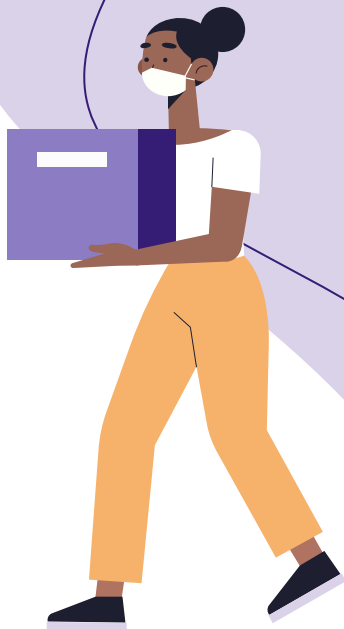
Object detection problem



			
Small YOLOv5s	Medium YOLOv5m	Large YOLOv5l	XLarge YOLOv5x
14 MB <sub>FP16</sub> 2.2 ms <sub>V100</sub> 36.8 mAP <sub>COCO</sub>	41 MB <sub>FP16</sub> 2.9 ms <sub>V100</sub> 44.5 mAP <sub>COCO</sub>	90 MB <sub>FP16</sub> 3.8 ms <sub>V100</sub> 48.1 mAP <sub>COCO</sub>	168 MB <sub>FP16</sub> 6.0 ms <sub>V100</sub> 50.1 mAP <sub>COCO</sub>

**02**

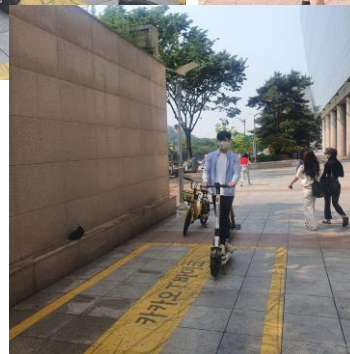
# **Dataset collection**





# Dataset Collection - photo taken by ourselves

Photo taken by ourselves - 103 sheets



# Dataset Collection - search on internet

Search on Internet - 205 + 213 (Additional) -> 423 sheets

You**Tube** Yandex Google





# Dataset Collection - Train / Valid

## Dataset Structure

```
| --yolov5
| --data
|   |--images      - 526 sheets
|       |--train   - 425 sheets
|       |--valid   - 101 sheets
|   |--labels      - 526 txt file
|       |--train   - 425 txt file
|       |--valid   - 101 txt file
```

Yolo V5 학습을 위한 폴더 트리

**Dataset - 526 sheets**

**Training Dataset - 425 sheets**

**Validation Dataset - 101 sheets**

# Dataset Labeling

## Dataset Labeling

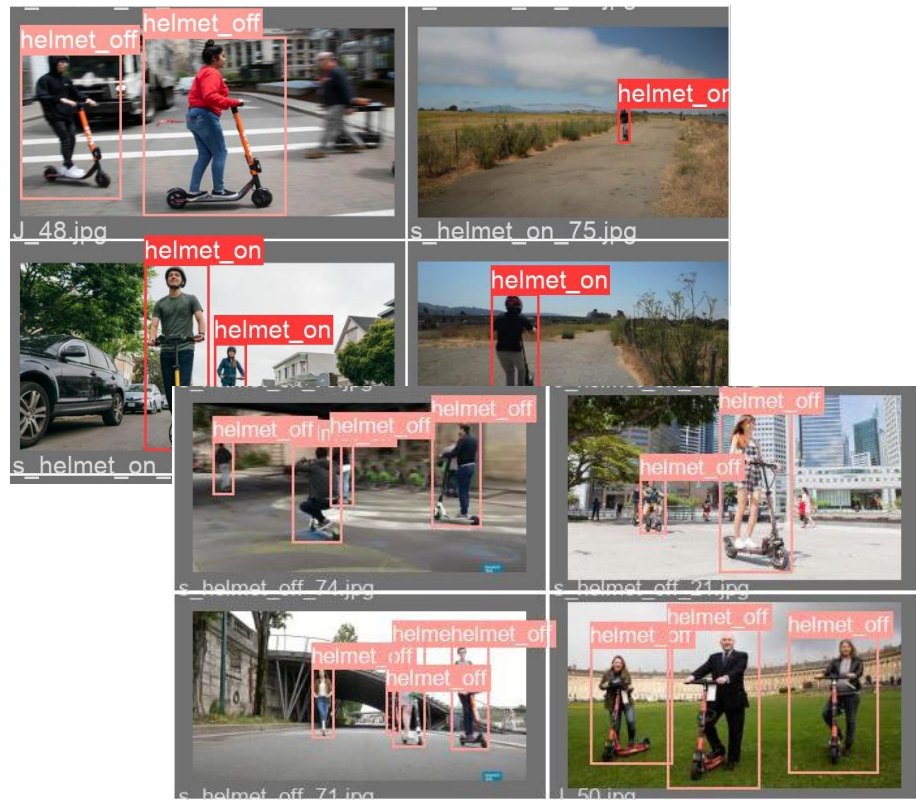
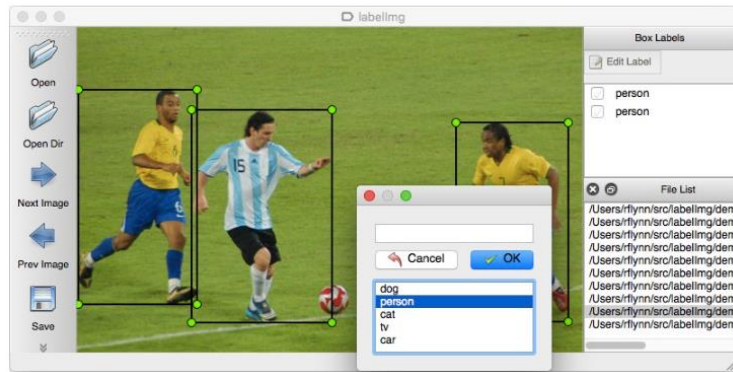
### Labellmg

pyl v1.8.6 build passing lang en lang zh  
lang zh-TW

Labellmg is a graphical image annotation tool.

It is written in Python and uses Qt for its graphical interface.


Annotations are saved as XML files in PASCAL VOC format, the format used by ImageNet. Besides, it also supports YOLO and CreateML formats.



# Dataset Collection - Augmentation

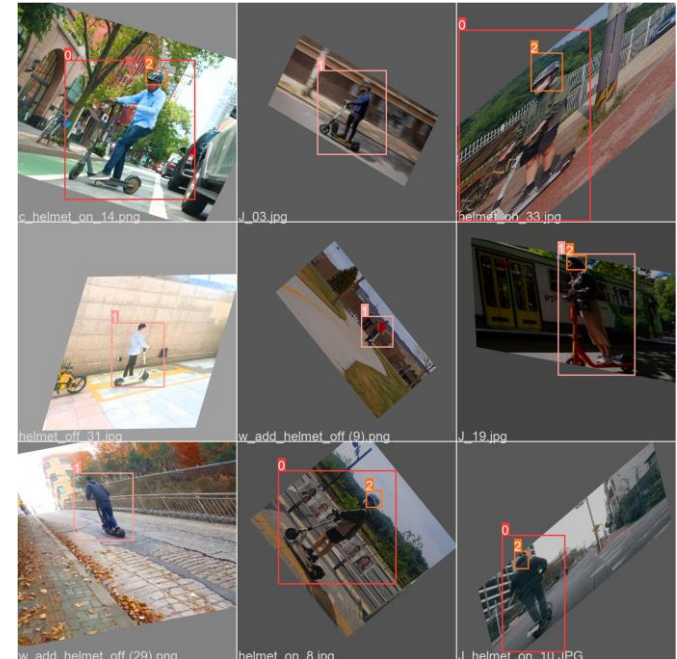
## Dataset Augmentation - Using yolov5 Albumentations

### Albumentations

 1.1.0 

Albumentations is a Python library for image augmentation. Image augmentation is used in deep learning and computer vision tasks to increase the quality of trained models. The purpose of image augmentation is to create new training samples from the existing data.

```
hsv_h: 0.015 # image HSV-Hue augmentation (fraction)
hsv_s: 0.7 # image HSV-Saturation augmentation (fraction)
hsv_v: 0.4 # image HSV-Value augmentation (fraction)
degrees: 0.0 # image rotation (+/- deg)
translate: 0.1 # image translation (+/- fraction)
scale: 0.5 # image scale (+/- gain)
shear: 0.0 # image shear (+/- deg)
perspective: 0.0 # image perspective (+/- fraction), range 0-0.001
flipud: 0.0 # image flip up-down (probability)
fliplr: 0.5 # image flip left-right (probability)
mosaic: 0.0 # image mosaic (probability)
mixup: 0.0 # image mixup (probability)
copy_paste: 0.0 # segment copy-paste (probability)
```



**03**

# **Transfer Learning**



# Problem : model don't know E-scooter



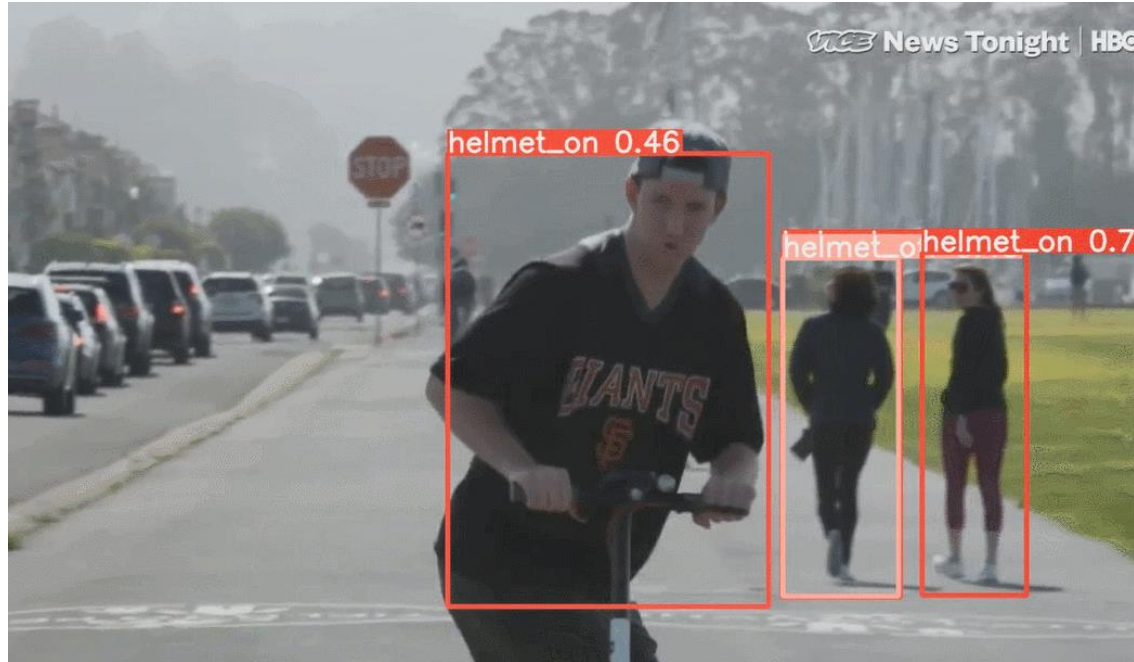
Image used for Object Detection



AI doesn't seem to detect e scooters,  
detected them as bicycle.



# Result - (Only) photo taken by ourselves



The lack of datasets, so the results of learning are very low



# Result - photo taken by ourselves + online data

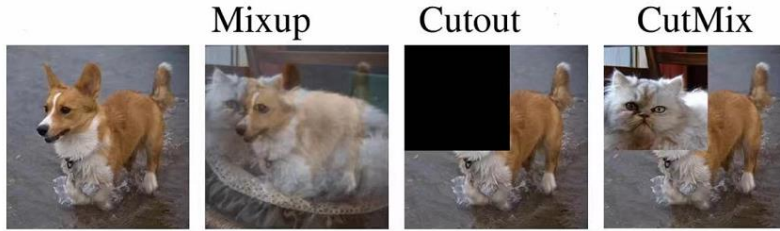


There was a problem that even people who did not ride the e scooter recognized it.

# Dataset Augmentation (Yolov5 Albumentation)

## Albumentation Parameter Tuning

Image



```
hsv_h: 0.015 # image HSV-Hue augmentation (fraction)
hsv_s: 0.7 # image HSV-Saturation augmentation (fraction)
hsv_v: 0.4 # image HSV-Value augmentation (fraction)
degrees: 0.0 # image rotation (+/- deg)
translate: 0.1 # image translation (+/- fraction)
scale: 0.5 # image scale (+/- gain)
shear: 0.0 # image shear (+/- deg)
perspective: 0.0 # image perspective (+/- fraction), range 0-0.001
flipud: 0.0 # image flip up-down (probability)
fliplr: 0.5 # image flip left-right (probability)
mosaic: 1.0 # image mosaic (probability)
mixup: 0.5 # image mixup (probability)
copy_paste: 0.0 # segment copy-paste (probability)
```



# Recognition Problem : Front, Rear



As datasets mainly related to the side were collected, problems related to this occurred due to low front, rear recognition rate.

# Collect Additional Datasets

Search on Internet - 205 + 213 (Additional) -> 423 sheets



# Improve the overall accuracy of the model

## Safety Helmet Wearing Detection Based on YOLOv5 of Attention Mechanism

Z P Xu, Y Zhang, J Cheng and G Ge

University of Jinan, Jinan, 250022, China

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**Abstract.** Aiming at problems of low accuracy and strong detection interference of the existing safety helmet wearing detection algorithms, an object detection algorithm by adding the squeeze-and-excitation block based on the YOLOv5 algorithm is proposed in this paper. The proposed method can not only obtain the weight of picture channel, but also accurately separate the foreground and background of the picture. Keeping all parameters unchanged, the proposed method and the YOLOv5 algorithm are applied to detect the safety helmet data set in the experiment. The result shows that the YOLOv5 algorithm with the squeeze-and-excitation block has an average detection accuracy of 94.5% for safety helmets and an average detection accuracy of 92.7% for human heads. The mAP value detected by the proposed method is 2% ~2.5% higher than using YOLOv5 algorithm directly.

In addition, we found a paper that adding a **SE layer** to an existing YOLOv5 model can increase the overall accuracy of the model.

After that, we added **SE layer** to the existing YOLOv5 model code.



# SE Layer Attention

```
class SELayer(nn.Module):
    def __init__(self, c1, r=16):
        super(SELayer, self).__init__()
        self.avgpool = nn.AdaptiveAvgPool2d(1)
        self.l1 = nn.Linear(c1, c1 // r, bias=False)
        self.relu = nn.ReLU(inplace=True)
        self.l2 = nn.Linear(c1 // r, c1, bias=False)
        self.sig = nn.Sigmoid()

    def forward(self, x):
        b, c, _, _ = x.size()
        y = self.avgpool(x).view(b, c)
        y = self.l1(y)
        y = self.relu(y)
        y = self.l2(y)
        y = self.sig(y)
        y = y.view(b, c, 1, 1)
        return x * y.expand_as(x)
```

Common.py

```
elif m is SELayer:
    channel = args[0]
    channel = make_divisible(channel * gw, 8) if channel != no else channel
    args = [channel]
```

yolo.py

```
# YOLOv5 v6.0 backbone
backbone:
  # [from, number, module, args]
  [[-1, 1, Conv, [64, 6, 2, 2]], # 0-P1/2
   [-1, 1, Conv, [128, 3, 2]], # 1-P2/4
   [-1, 3, C3, [128]],
   [-1, 1, Conv, [256, 3, 2]], # 3-P3/8
   [-1, 6, C3, [256]],
   [-1, 1, Conv, [512, 3, 2]], # 5-P4/16
   [-1, 9, C3, [512]],
   [-1, 1, Conv, [1024, 3, 2]], # 7-P5/32
   [-1, 3, C3, [1024]],
   [-1, 1, SPPF, [1024, 5]], # 9
   [-1, 1, SELayer, [1024, 4]]
```

yolov5x.yaml





# 04

## Evaluation & Analysis

# Github & Wandb



**Wandb**

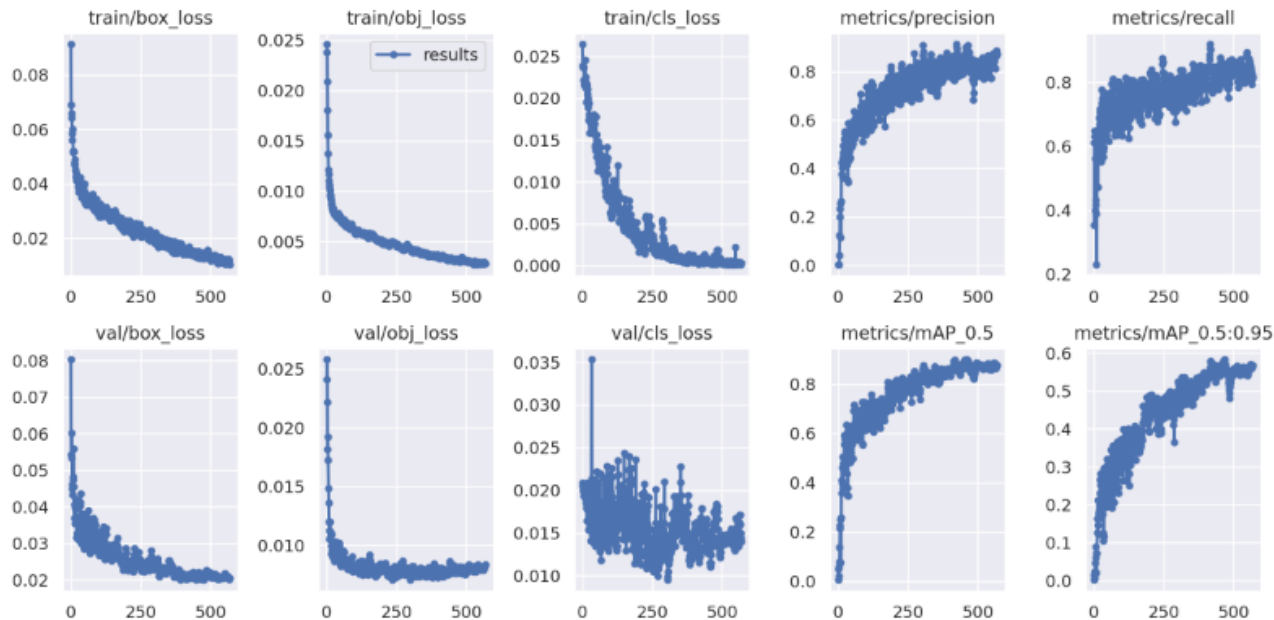
[Weights & Biases \(wandb.ai\)](https://wandb.ai)



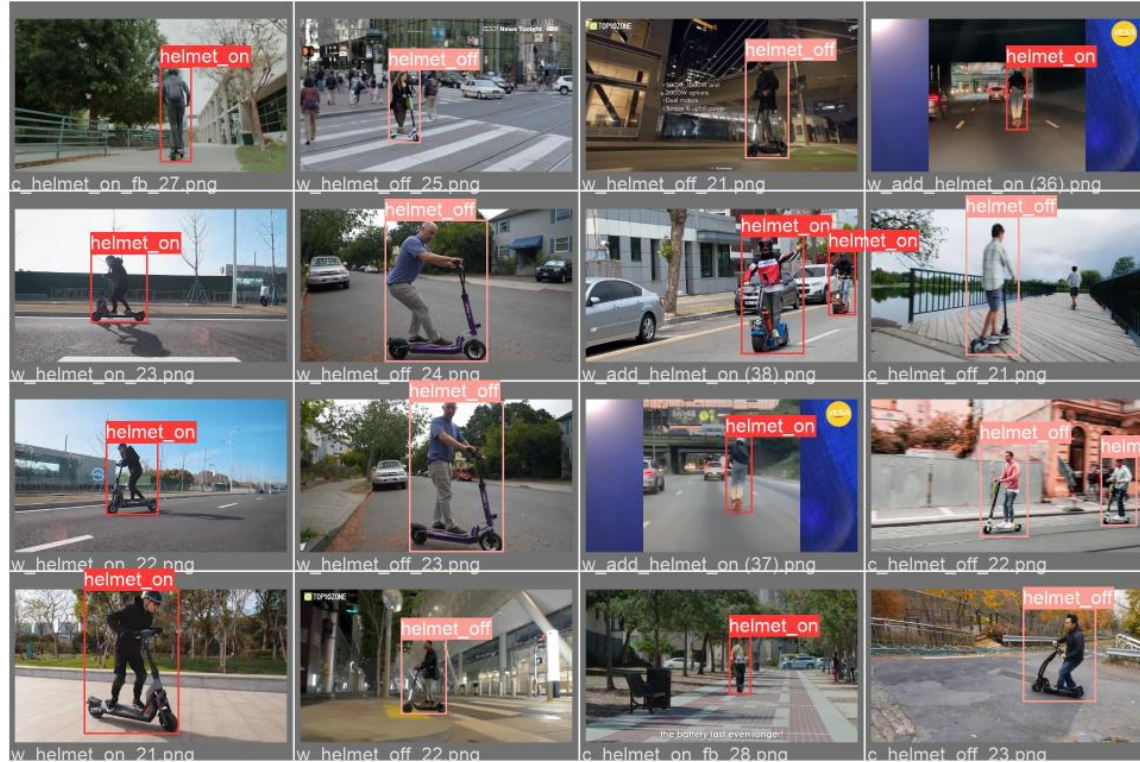
**Github**

[https://github.com/dntjr41/DeepLearning\\_TermP](https://github.com/dntjr41/DeepLearning_TermP)

# Train / Valid Result Graphs



# Validation results



# Result on Transfer learning



Before training, the trained model doesn't detect



At final-model, AI Now can find human with helmet-unequipped after Testing with custom dataset.



# Result on training : Final model



AI understands what is e scooter  
Also AI can determine if Rider equipped helmet or not

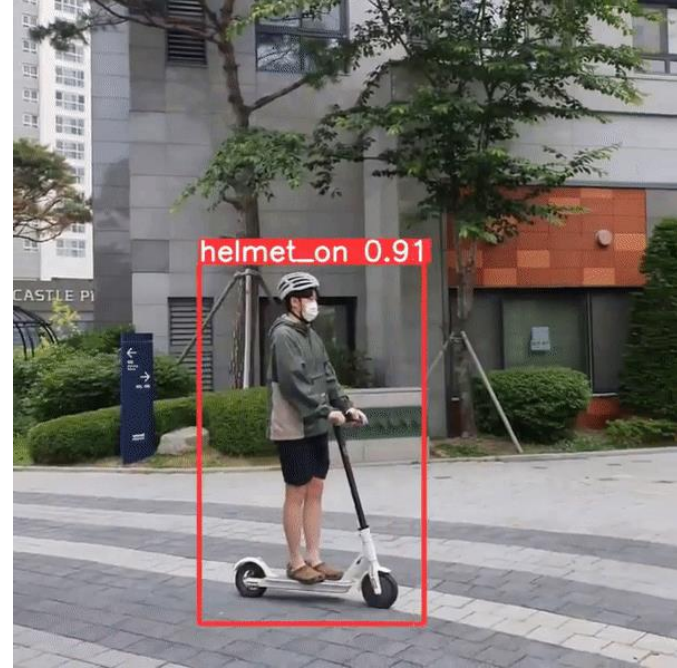


# Result on training : Final model



AI understands what is e scooter  
Also AI can determine if Rider equipped helmet or not

# Result on training : Final model



Result of video taken by ourselves

# Self- Review on project : Feedback after project

## 권순형

- Through our experience in learning deep learning models, we realize that the right datasets and appropriate augmentation have a great influence on the results.
- I realized that deep learning can be used in a wide variety of ways.

## 심우석

- In order to improve overall accuracy, it would have been better to add a background to the dataset.
- It was difficult to find the right parameter for learning, but we liked the process of correcting the results by checking that they were getting better.

## 오찬희

- Through the project, We were able to analyze the results, add data sets, and expand performance through model attention to improve skills and understanding.
- I think it will be possible to improve performance enough with more data sets and time, and I hope that this improvement will contribute greatly to the detection of the electric scooter helmet.

## 전강훈

- Finding using studied Knowledge from this semester into project was resulted positive experience to me.
- Also the result was good, but though what if i choiced different methods such as difference on layers, or models, or other method and techniques to gain best answer.

# 05

## Team Introduction



# Team Introduction



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# References

## Websites

- Tzutalin , labellmg project, <https://github.com/tzutalin/labellmg#usage>, updated on 2022.04.30, accessed on 2022.05.22
  - Albumentations-team , albumentations , <https://github.com/albumentations-team/albumentations>, updated on 2022.02.13, accessed on 2022.05.22
  - Squeeze-and-Excitation Networks, <https://arxiv.org/pdf/1709.01507.pdf>
  - Ultralytics, YOLO v5 repository, <https://github.com/ultralytics/yolov5>, last updated on 2022.05.22, accessed on 2022.05.22
  - Wandb, <https://wandb.ai/site>
  - Yolov5 tutorial, <https://sguys99.github.io/ds01>
  - <https://iopscience.iop.org/article/10.1088/1742-6596/2213/1/012038>
- 



# THANKS

Do you have any questions?

Team B -

심우석, 전강훈, 오찬희, 권순형

